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Converting a Converging Algorithm into a

Polynomially Bounded Algorithm

by George B. Dantzig

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Abstract: We consider the general Phase I linear programming problem with a convexity constraint which can be written after some algebraic manipulation in the form:

Find
$$x_j \ge 0$$
, $\sum_{1}^{n} P_j x_j = 0$, $\sum_{1}^{n} x_j = 1$

where P_j are m-vectors satisfying $||P_j|| = 1$. If feasible, von Neumann's Center of Gravity Algorithm generates a sequence $t = 1, 2, \ldots$ of approximate solutions $\sum P_j x_j^t = b^t$, $\sum x_j^t = 1$, $x_j^t \geq 0$ which converges in the limit as $t \to \infty$ to a feasible solution to the Phase I problem. We assume that all perturbed problems $\sum_{1}^{n} P_j x_j = \hat{b}$, $\sum x_j = 1$, $x_j \geq 0$ are feasible for all $||\hat{b}|| < r$ where r > 0 is given. We apply this algorithm to m+1 perturbed problems with right hand sides $\hat{b} = \hat{b}^i$, $i = 1, 2, \ldots, m+1$ to obtain an exact solution to the unperturbed problem with $\hat{b} = 0$ in $T < 4r^{-2}(m+1)^3$ iterations. Each iteration consists of $m(n+3)\delta$ multiplications and additions where δ is the non-zero coefficient density.

Von Neumann* in 1948 proposed the first interior algorithm for solving a general Phase I linear program with a convexity constraint. We will reproduce his proof that in $t < 1/\rho^2$ iterations an approximate solution $\sum P_j x_j^t = b^t$ will be generated with $||b^t|| < \rho$. When applied to a perturbed problem $b = \hat{b} \neq 0$, we will show that in $t < 4/\rho^2$ iterations an approximate solution will be generated with $||b^t - \hat{b}|| < \rho$.

^{*} verbal communication

Geometrically, in the m-space of the columns, since $\|P_j\|=1$, all points P_j lie on the surface of the m-dimensional hypersphere S_0 of unit radius with center at the origin. We are given r the radius of a concentric hypersphere $S_1 \subseteq S_0$ centered at the origin that lies in the convex hull of the points P_j . Thus r is a measure of how deeply the origin is embedded in the set of b such that $b = \sum P_j x_j$, $x_j \ge 0$, $\sum x_j = 1$ is feasible.

To generate the m+1 different finite sequences (x^t, b^t) whose b^t approach m+1 different points \hat{b}^i , the \hat{b}^i are prechosen. These can be the vertices of any simplex lying in the set of feasible b that contains the origin as an interior point. We choose \hat{b}^i to be the vertices of an (m+1) equilateral simplex whose center is the origin and whose vertices are located at distances $r \cdot m/(m+1)$ from the origin; for example the coordinates of \hat{b}^i may be chosen as follows:

where $a_i = r \sqrt{\frac{m}{m+1}} \ \cdot \sqrt{\frac{1}{i(i+1)}}$.

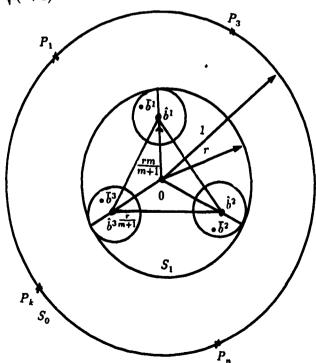


Figure 1. The Iterations Converge to \hat{b}^i Instead of the Origin 0.

When the i^{th} sequence (x^t, b^t) (which is converging towards \hat{b}^i) reaches a point $b^t = \bar{b}^i$ such that $\|\bar{b}^i - \hat{b}^i\| < r/(m+1)$, the sequence for that i is terminated. Note that all interior points of Ball_i of radius $\rho = r/(m+1)$ centered at \hat{b}^i lie inside the hypersphere $S_1 \subseteq S_0$. We will show $b^t = \bar{b}^i \subset \text{Ball}_i$ is attainable by the iterative process. Associated with \bar{b}^i is the approximate solution $\bar{x}^i = x^t$ that generated it. Thus an upper bound to generate all m+1 approximate solutions (\bar{x}^i, \bar{b}^i) whose \bar{b}^i lie strictly in m+1 ρ -balls centered at \hat{b}^i can be done in

(2) iteration count
$$< 4(m+1)/\rho^2 = 4(m+1)^3/r^2$$
, $\rho = r/(m+1)$,

iterations. The final step is to generate the feasible solution \bar{x} to the Phase I problem by finding weights $\bar{\lambda}_i > 0$, $\bar{x} = \sum \lambda_i \bar{x}^i \ge 0$, $\sum \bar{x}_j = 1$, 1, $\sum P_j \bar{x}_j = 0$. These weights $\bar{\lambda} = (\bar{\lambda}_1, \bar{\lambda}_2, \dots, \bar{\lambda}_{m+1})$ are found by solving the $(m+1) \times (m+1)$ system

(3)
$$\sum \bar{b}^i \bar{\lambda}_i = 0$$

$$\sum \bar{\lambda}_i = 1.$$

We will prove that this system has a unique solution $\bar{\lambda} = (\bar{\lambda}_1, \dots, \bar{\lambda}_{m+1}) > 0$.

We now describe the detailed steps of von Neumann's algorithm for finding an approximate solution to a perturbed problem $\sum P_j x_j = \hat{b}$, $\sum x_j = 1$, $x \ge 0$ and give a proof of the rate of convergence of the *i*-th sequence to some $\hat{b} = \overline{b}^i \subset B_i$. We initiate the sequence of iterations by $x = x^1 = (1, 0, ..., 0)$, $b^1 = P_1$. Inductively let x^{t-1} , b^{t-1} be the t-1 approximation. We use it to generate x^t , b^t .

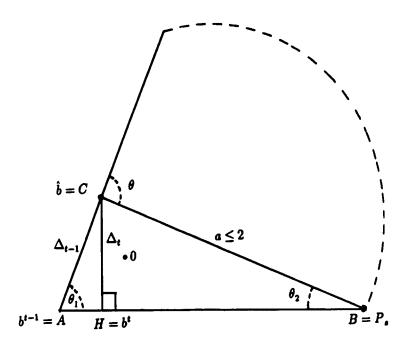


Figure 2. The Von Neumann Iterative Step

Referring to Figure 2, P_s is selected as that P_j such that $P_j - \hat{b}$ makes the sharpest angle θ with direction $\hat{b} - b^{t-1}$, namely

(4)
$$s = \text{ARGMAX } |[\hat{b} - b^{t-1}]^{\mathsf{T}} [P_j - \hat{b}] / ||P_j - \hat{b}|| |.$$

which can be carried out in m(n+3) operations assuming $\|P_j - \hat{b}\|$ is preprocessed. The triangle b^{t-1} , P_s , \hat{b} will be labeled ABC. The next approximation point $H = b^t$ is the foot of perpendicular dropped from C onto the side AB of the triangle ABC. From the figure, it is clear that H is a weighted convex combination of A and B with weights proportional to $\cos \theta_2$ and $\cos \theta_1$, i.e.,

(5)
$$b^{t} = (\cos \theta_{2} \cdot b^{t-1} + \cos \theta_{1} \cdot P_{s})/(\cos \theta_{2} + \cos \theta_{1}),$$

$$x^{t} = (\cos \theta_{2} \cdot x^{t-1} + \cos \theta_{1} \cdot U_{s})/(\cos \theta_{2} + \cos \theta_{2}),$$

where U_s is the unit n vector with 1 in component s. $\cos \theta_1$ and $\cos \theta_2$ are computed by

(6)
$$\cos \theta_2 = \frac{(\hat{b} - P_s)^{\mathrm{T}} (b^{t-1} - P_s)}{\|\hat{b} - P_s\| \|b^{t-1} - P_s\|}, \quad \cos \theta_1 = \frac{(P_s - b^{t-1})^{\mathrm{T}} (\hat{b} - b^{t-1})}{\|P_s - b^{t-1}\| \|\hat{b} - b^{t-1}\|}.$$

In order to determine the rate of convergence, note $\theta \leq \pi/2$ because if, on the contrary, $\theta > \pi/2$ then all points P_j would lie on one side of the hyperplane through \hat{b} orthogonal to $b^{t-1} - \hat{b}$ implying that $\hat{b} = \hat{b}^i$ for the *i*-th sequence lies outside the convex hull of the P_j 's contrary to our assumption that all points located at a distance r or less from the origin are in the set of feasible b (i.e., \hat{b}^i by construction lies in the interior of the set of feasible $\hat{b} \subset S_1$ at a distance r/(m+1) from the boundary of S_1 . To simplify the notation, let

$$\Delta_{t-1} = ||b^{t-1} - \hat{b}|| \text{ and } \Delta_t = ||b^t - \hat{b}||,$$

then

(7)
$$\Delta_t = \Delta_{t-1} \sin \theta_1 \text{ and } \Delta_t = ||P_s - \hat{b}|| \sin \theta_2.$$

Therefore, noting $\theta_1 + \theta_2 = \theta \le \pi/2$,

$$\left(\frac{\Delta_t}{\Delta_{t-1}}\right)^2 + \left(\frac{\Delta_t}{\|P_s - \hat{b}\|}\right)^2 = \sin^2\theta_1 + \sin^2\theta_2 \le 1.$$

Recalling that diameter of the hypersphere is 2, it follows that $||P_s - \hat{b}|| < 2$ and therefore for $\tau = 2, 3, ..., t$:

(8)
$$\left(\frac{\Delta_{\tau}}{\Delta_{\tau-1}}\right)^2 + \left(\frac{\Delta_{\tau}}{2}\right)^2 < 1.$$

Comment: These inequalities can be made tighter when $\hat{b}=0$ because $\|P_s - \hat{b}\| = \|P_s\| = 1$. If so, (8) can be replaced by $(\Delta_\tau/\Delta_{\tau-1})^2 + \Delta_\tau^2 \le 1$ and the development that follows can be modified accordingly with the conclusion that if the von Neumann iterative process is applied to the case $\hat{b}=0$ instead of to $\hat{b}^i \ne 0$ an approximation b^i such that $\|b^i\| < \rho$ can be attained in less that $1/\rho^2$ iterations (instead of less than $4/\rho^2$ iterations).

Dividing (8) through by $(\Delta_{\tau})^2$ for $\tau = 2, ..., t$:

$$(1/\Delta_{t-1})^2 + (1/4) < (1/\Delta_t)^2$$

$$(1/\Delta_{t-2})^2 + (1/4) < (1/\Delta_{t-1})^2$$

$$\vdots \qquad \vdots \qquad \vdots$$

$$(1/\Delta)^2 + (1/4) < (1/\Delta_2)^2.$$

Summing the above, canceling terms common to both sides of the sum and, recalling $\Delta_1 < 2$, we have

(10)
$$(1/\Delta_t)^2 > (1/4) + (t-1)/4 = t/4.$$

We conclude that $t < 4/\Delta_t^2$ iterations, i.e. less than $4/\rho^2$ iterations would be needed for the i^{th} sequence to terminate by reaching $b^t = \bar{b}^i$, an interior point of the ρ -ball centered at \hat{b}^i . Since $\rho = r/(m+1)$ and there are (m+1) ρ -balls, the upper bound on

(11) iteration count
$$< 4(m+1)^3/r^2$$
.

What remains to show is that the $(m+1)\times (m+1)$ system (3) can be solved, that the solution $\bar{\lambda}$ is unique, and that $\hat{\lambda} = (\bar{\lambda}_1, \bar{\lambda}_2, \dots, \bar{\lambda}_{m+1}) > 0$.

Existence of Separating Hyperplanes: Let $y = (y_1, y_2, ..., y_m)$ represent a general point in \mathbb{R}^m . The equation of any hyperplane through the origin has the form $a^Ty = 0$. This hyperplane is said to separate y^1 from y^2 if a^Ty^1 and a^Ty^2 are of opposite signs.

Fact 1. Each hyperplane $(\hat{b}^i)^T y = 0$ for i = 1, 2, ..., m separates any point in the ρ -ball centered at \hat{b}^i from any point lying in any of the other ρ -balls centered at \hat{b}^j .

Proof: Because of the m+1 fold symmetry of the equilateral simplex it is sufficient to demonstrate that the hyperplane $(\hat{b}^{m+1})^{\mathrm{T}}y=0$ separates \overline{b}^{m+1} from \overline{b}^m where $\|\overline{b}^{m+1}-\hat{b}^{m+1}\| < r/(m+1)$ and $\|\overline{b}^m-\hat{b}^m\| < r/(m+1)$. The coordinates of \hat{b}^{m+1} and \hat{b}^m defined by (1) are $\hat{b}^{m+1}=(0,0,\ldots,rm/(m+1))^{\mathrm{T}}$ and $\hat{b}^m=(0,0,\ldots,r\sqrt{m-1}/\sqrt{m+1},-r/(m+1))^{\mathrm{T}}$. The hyperplane $(\hat{b}^{m+1})^{\mathrm{T}}y=0$ reduces to $(0,\ldots,1)y=U_m^{\mathrm{T}}y=0$. Letting

The Separating Hyperplanes Theorem below states conditions which imply that the points $\bar{b}^1, \bar{b}^2, \dots, \bar{b}^{m+1}$ are the vertices of a simplex containing the origin in its interior. That these conditions are satisfied follows from Fact 1.

Separating Hyperplanes Theorem: Given (1) that $(\hat{b}^1, \hat{b}^2, ..., \hat{b}^{m+1})$ are any (m+1) vertices of an m-dimensional simplex \hat{T} containing the origin; given (2) that $a^iy = 0$ for i = 1, 2, ..., m+1 are the equations of m+1 hyperplanes separating \hat{b}^i from \hat{b}^j for all $j \neq i$; and given (3) any m+1 points $\bar{b}^1, \bar{b}^2, ..., \bar{b}^{m+1}$ such that each hyperplane $a^iy = 0$ separates \bar{b}^i (on the same side as \hat{b}^i) from \hat{b}^j for all $j \neq i$; then $\bar{b}^1, \bar{b}^2, ..., \bar{b}^m$ are the vertices \bar{T} of an m-dimensional simplex that contains the origin as an interior point.

Proof: Since the simplex associated with \hat{T} contains the origin, we know there exist $\hat{\lambda}_i \geq 0, \overline{\lambda}_i \geq 0$ such that

$$(13.1) \qquad \qquad \sum \hat{b}^{j} \hat{\lambda}_{j} + \sum \overline{b}^{i} \overline{\lambda}_{i} = 0$$

Before continuing with the proof, we show two more facts:

Fact 2. If $(\hat{\lambda}, \bar{\lambda})$ is a feasible solution to (13.1), (13.2), then $\hat{\lambda}_i + \bar{\lambda}_i > 0$ for all i.

Suppose, on the contrary, $\hat{\lambda}_k = 0$, $\bar{\lambda}_k = 0$ for some k. Multiply (13.1) on the left by a^k ; recall, by assumption, $a^k \hat{b}^j < 0$ and $a^k \bar{b}^j < 0$ for all $j \neq k$. We have

(14.1)
$$\sum_{1 \neq k} (a^k \hat{b}^i) \hat{\lambda}_i + \sum_{j \neq k} (a^k \bar{b}^i) \bar{\lambda}_i = 0$$

implying, that (14.1) is the sum of non-negative terms (not all zero by (14.2), a contradiction.

Fact 3. If T is any simplex containing the origin whose vertices i are separated from the remaining vertices $j \neq i$ by a hyperplane $a^i y = 0$ for each i, then T contains the origin strictly in its interior.

Fact 3 follows from Fact 2 by setting $\bar{b}^i = \hat{b}^i$ for all i.

Continuing with the proof of the separating hyperplanes theorem, define ${\mathfrak B}$ and U_{m+1} by

(15)
$$\mathfrak{B} = \begin{bmatrix} \hat{b}^1 & \hat{b}^2 & \dots & \hat{b}^{m+1} \\ 1 & 1 & & 1 \end{bmatrix}, \qquad U_{m+1} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

Since \hat{T} are the vertices of an m-dimensional simplex by assumption, it means that \mathfrak{B} is non-singular and that $\mathfrak{B}\hat{\lambda} = U_{m+1}$ can be solved for $\hat{\lambda}$ and, when solved, $\hat{\lambda} \geq 0$. From Fact 3 it follows that $\hat{\lambda} > 0$. We view \mathfrak{B} as a feasible non-degenerate basis and consider $\begin{bmatrix} \bar{b}^1 \\ 1 \end{bmatrix}$ as an incoming non-basic column. We assert it will replace $\begin{bmatrix} \hat{b}^1 \\ 1 \end{bmatrix}$ in the basis because, on the contrary, if it replaced some column $k \neq 1$ in the basis, it would imply after the replacement that both $\bar{\lambda}_k$ and $\hat{\lambda}_k$ are 0 in a feasible solution, contrary to Fact 2. By replacing in turn basis columns $\begin{bmatrix} \hat{b}^2 \\ 1 \end{bmatrix}$ by $\begin{bmatrix} \hat{b}^2 \\ 1 \end{bmatrix}$, $\begin{bmatrix} \hat{b}^3 \\ 1 \end{bmatrix}$ by $\begin{bmatrix} \bar{b}^3 \\ 1 \end{bmatrix}$, etc., we arrive at the conclusion that \bar{T} are the vertices of a simplex containing the origin. It then follows from Fact 3 that this simplex contains the origin as a strictly interior point.

This completes the proof that the (m+1) sequences converge to m+1 points \bar{b}^i in less than $4(m+1)^3/r^2$ iterations. By applying the weights $\bar{\lambda}_i > 0$ to the corresponding \bar{x}^i , we generate the exact solution x to the Phase I linear program.

One final remark: Just because an algorithm is polynomial does not necessarily make it practical. The von Neumann algorithm has a poor convergence rate. Like the simplex method each of its iterations requires about $mn\delta$ multiplications and additions where δ is the density of non-zero coefficients. When applied to (m+1) perturbed problems as we do in this paper, we obtain an upper bound of $4(m+1)^3/r^2$ iterations where 0 < r < 1. The moral of this tale is that, like gunners, we may do better by first bracketing the target and then applying a final correction.

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where P_j are m-vectors satisfying $||P_j||=1$. If feasible, von Newmann's Center of Gravity Algorithm generates a sequence $t=1,2,\ldots$ of approximate solutions $\sum P_j x_j^t = b^t, \ \sum x_j^t = 1, \ x_j^t \geq 0$ which converges in the limit as $t \to \infty$ to a feasible solution to the Phase I problem. We assume that all perturbed problems $\sum_{1}^{n}P_{j}x_{j}=\hat{b},\;\sum x_{j}=1,\;x_{j}\geq0$ are feasible for all $\parallel\hat{b}\parallel< r$ where r>0is given. We apply this algorithm to m+1 perturbed problems with right hand sides $\hat{b}=\hat{b}^i,\ i=1$ $1,2,\ldots,m+1$ to obtain an exact solution to the unperturbed problem with $\hat{b}=0$ in $T<4r^{-2}(m+1)^3$ iterations. Each iteration consists of $m(n+3)\delta$ multiplications and additions where δ is the non-zero coefficient density.

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